

NAME: SHAILJA MISHRA

ROLL NO. 1918081

SECTION: D4

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Assignment-1

1. Asymptotic Notations are methods / languages using which we can define the running time of algorithm based on input size.

To represent the upper & lower bounds, we need some kind of syntax & this is represented in form of functional $f(n)$.

→ logarithmic $\rightarrow \log n$

Linear $\rightarrow n$

→ Quadratic $\rightarrow n^2$

polynomial $\rightarrow n^2$

→ Exponential $\rightarrow a^n$

2. for $(i=1 \text{ to } n) \{ i = i * 2 \}$
'i' is doubling everytime.

for k^{th} step $\rightarrow 2^k \leq n$ & for $(k+1)$ we are out of loop
Taking log both sides

$$\log 2^k \leq \log n$$

$$k \leq \log_2 n$$

$$\text{Time complexity} \leq O(\log n)$$

3. $T(n) \leq 3T(n-1)$ if $n > 0$ otherwise 1
 $T(n) \leq aT(n-b) + f(n)$ (Master theorem)
 $a \geq 3, b \geq 1$

$$f(n) = 0, k \geq 0$$

$$T(n) \leq O(n^k a^{1/2})$$

$$T(n) \leq O(n a^n)$$

$$T(n) \leq O(3^n)$$



④ $T(n) = 2T(n-1) - 1$
 $T(n) = aT(n-b) + f(n)$

$a=2$

$b=1$

$T(n-1) = 2T(n-2) - 1$

$T(n) = 2(2T(n-2) - 1) - 1$
 $= 4T(n-2) - 3$

Similarly,

$T(n) = 8T(n-3) - 7$

$T(n) = 2^k T(n-k) - (2^k - 1)$

Let $n-k=1$

$T(n) = 2^k T(1) - 2^k + 1$

$= 2^k (T(1) - 1) + 1$

$\Rightarrow T(n) = O(2^k) = O(2^n) \text{ Ans.}$

⑤

```
int i = 1, s = 1;
while (s <= n) {
    i++; s = s + i;
    printf("#");
}
```

We can see that 's' is increasing by 1.
 $\therefore O(n) \text{ Ans.}$

⑥

```
void function (int n) {
    int i, count = 0;
    for (i = 1; i * i <= n; i++)
        count++;
}
```

$n = 5$

$i = 1 \quad 1 \times 1 \leq n$

$i = 2 \quad 2 \times 2 \leq n$

$i = 3 \quad \times \text{ out of loop.}$

Loop is working for $n/2$ time only.
 $\therefore O(n/2) \Rightarrow O(n) \text{ Ans.}$

⑦. void function (int n) {
 int i, j, k, count = 0;
 for (i = n/2; i <= n; i++)
 for (j = 1; j <= n; j = j * 2)
 for (k = 1; k <= n; k = k * 2)
 count++;
 i loop = $O(n/2)$
 j loop = $O(\log n)$
 k loop = $O(\log n)$

∴ Final = $O(n \log^2 n)$ Ans.

⑧. fun (int n) {
 if (n == 1) return;
 for (i = 1 to n) {
 for (j = 1 to n) {
 print (n);
 }
 }
 fun (n - 3);
 }

Time Complexity $T(n) = T(n^2) - 3$ Ans.

⑨. void func (int n) {
 for (i = 1 to n) → $O(n)$
 for (j = 1; j <= n; j = j + i) → $O(\log n)$
 printf ('n');
 ∴ $O(n \log n)$ Ans.

⑩. $f(n) = n^k$ $k \geq 1$
 $g(n) = a^n$ $a > 1$

$f(n) = O(g(n))$
 $n^k = O(a^n)$

Take log

$k \log n = n \log a$

$\frac{\log n}{\log a} = \frac{n}{k}$

$\log \frac{n}{a} = \frac{1}{k} n$

Let $\frac{1}{k} = C$

$\therefore O(Cn)$ is time complexity.

⑪ void fun (int n) { int j=1, i=0;
while (i < n) { i = i+j; j++; } }

→

i	j	loop runned
0	1	1
1	2	2
3	3	3

We can observe that loop is running for $n/2 + 1$ times.

$\therefore O(n/2 + 1)$
 $= O(n)$ Ans.

⑫. `int fibn(int n) {`
 line 1 `if (n==0 || n==1) return n;`
 line 2 `else return fib(n-1) + fib(n-2);`

We know that line 1 takes $O(1)$ time while
 Line 2 takes $T(n-1) + T(n-2)$
 \therefore recursive eqⁿ = $T(n-1) + T(n-2) + O(1)$

$$\therefore T(n) = T(n-1) + T(n-2) \quad - (1)$$

$$T(n-1) = T(n-2) + T(n-3) \quad - (2)$$

$$T(n) = T(n-2) + T(n-2) + T(n-3) \quad - (3)$$

$$T(n-2) = T(n-3) + T(n-4) \quad - (4)$$

$$T(n) = 2T(n-3) + T(n-4)$$

$$\therefore T(n) = 2T(n-k) + T(n-(k+1))$$

$$\boxed{O(n) = 2^n}$$

$$\boxed{\text{Space Complexity} = O(n)} \quad \text{Ans}$$

⑬ (i) $(n \log n)$

```
→ void flntn() {
    for (i=0; i<n; i++)
        for (j=0; j<n; j=j*2)
            count++;
}
```

(ii) n^3

```
for (i=0; i<=n; i++)
    for (j=0; j<n; j++)
        for (k=0; k<n; k++)
        {
            printf("GEU")
        }
```


(iii) $\log(\log n)$
for ($i=0; i < \log n; i = i * 2$)
printf("Hi");

(14) $T(n) = T(n/4) + T(n/2) + Cn^2$
Using Master's Theorem
We know that ; $T(n/2) \geq T(n/4)$
 $\therefore T(n) \leq 2T(n/2) + Cn^2$
Apply Master's Theorem to RHS

$$T(n) \leq O(n^2)$$

$$T(n) = O(n^2)$$

Also,

$$T(n) \geq Cn^2$$

$$T(n) = O(n^2)$$

$$T(n) \Omega(n^2)$$

Since

$$T(n) = O(n^2)$$

$$T(n) = \Omega(n^2)$$

$$\therefore T(n) = O(n^2) \text{ Ans.}$$

(15) int fun(int n) {
for ($i=1; i \leq n; i++$)
for ($j=1; j \leq n; j=j+1$) { $O(1)$ }

for the i loop = $O(n)$
for j loop = $O(\log n)$

$$\therefore O(n \log n) \text{ Ans.}$$

(16) for ($i=2; i \leq n; i = \text{pow}(i, k)$)
{ $O(1)$ }

$$n=7$$

$$k=2$$

$$\rightarrow \left. \begin{matrix} i \\ 2 \\ 4 \\ 16 \end{matrix} \right\} \begin{matrix} 1 \\ 2 \\ x \end{matrix} \quad \left. \begin{matrix} 2, 2^k, (2^k)^k, (2^k)^{k^2} \end{matrix} \right\}$$

$$2^{k \log k \log(n)} = O(\log(\log n)) \text{ Ans.}$$

①7.

Quick Sort

When quick sort repeatedly divides the array into two parts of 99% and 1%, it is the worst case.

Recurrence Relation

Let current problem = n

" $(n-1)$ " we have to call

Quicksort Worst Case

$$T(n) = T(n-1) + T(1) + C \cdot n$$

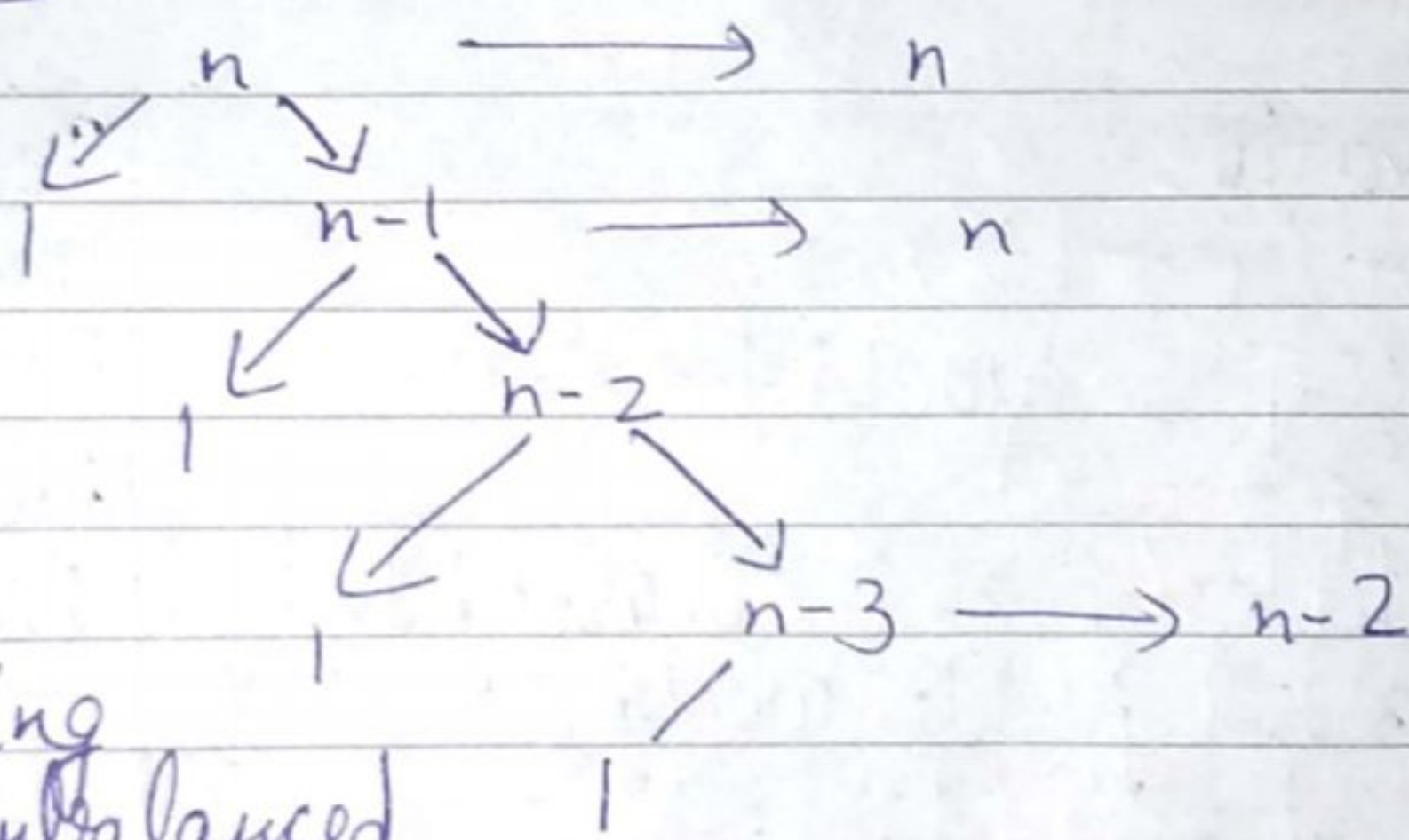
$$T(n) = T(n-1) + \Theta(1) + C \cdot n$$

$$= \sum_{k=1}^n \Theta(k)$$

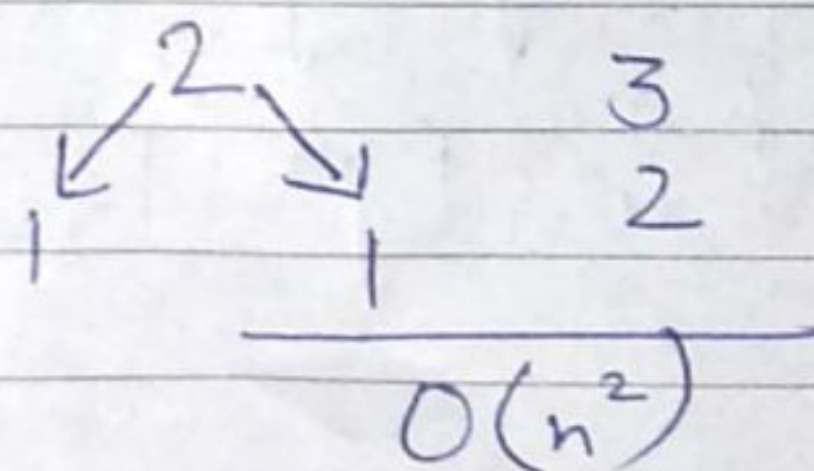
$$= \Theta\left(\sum_{k=1}^n k\right)$$

$= \Theta(n^2)$ is the time complexity.

Recursion Tree:



If the partitioning is maximally unbalanced at every recursive step of algorithm.





(18). a) $100 < \log \log n < \log n < n \log n < \text{root}(n) < \log n < n$
 $< \sqrt[n]{n} < 2^n < 2^{2n} < 4^n < n!$

(19). void linearSearch (int arr[], int n, int x)
 {
 if (arr[n/2] == x) {
 for (i = 0 to n/2 + 1)
 {
 if (arr[i] == x)
 {
 cout << "found";
 break;
 }
 }
 }
 else {
 for (i = n/2 + 1; i < n; i++)
 {
 if (arr[i] == x)
 {
 cout << "found";
 break;
 }
 }
 }
 }

Iterative	Recursive
<pre> for (i = 1 to n-1) Key ← arr[i] j = i - 1 while (j >= 0 && arr[j] > Key) arr[j+1] = arr[j] j = j - 1; arr[j+1] ← Key; </pre>	<pre> void insertion (arr, n) { if (n <= 1) return; insertion (arr, n-1); last ← arr[n-1]; j ← n-2; while (j >= 0 && arr[j] > last) { arr[j+1] ← arr[j] j = j - 1; } arr[j+1] = last; } </pre>

It is known as online sort because it does not need to know anything about values, it will sort & the info is requested WHILE the algo is running
 # It gets new value at every iteration. Examples → Selection & Insertion



21. Bubble Sort
Best = $O(n)$
Worst = $O(n^2)$
Avg = $O(n^2)$

Selection Sort
Best = $O(n^2)$
Worst = $O(n^2)$
Avg = $O(n^2)$

Quick Sort
Best = $O(n \log n)$
Worst = $O(n^2)$
Avg = $O(n \log n)$

Insertion Sort
Best = $O(n)$
Worst = $O(n^2)$
Avg = $O(n^2)$

Merge Sort
Best = $O(n \log n)$
Worst = $O(n \log n)$
Avg = $O(n \log n)$

22. In Place
Bubble
Insertion
Selection
Quick, Heap
* Not Inplace
Merge

Stable
Insertion
Merge
Bubble
* Not Stable
Quick, Heap

Online
Selection
Insertion
* Offline
Bubble, Quick,
Merge

23. Iterative Binary Search

```
int binary_search(int arr[], int l, int r, int x)
{
    while (l <= r) {
        int mid = (l+r)/2;
        if (arr[mid] == x)
            return mid;
        if (arr[mid] < x)
            l = mid + 1;
        else
            r = mid - 1;
    }
    return -1;
}
```




Recursive Binary Search

```
int binarySearch(int arr[], int l, int r, int x)
{
    while (l <= r) {
        int mid = (l+r)/2;
        if (arr[mid] == x) return mid;
        else if (arr[mid] > x)
            return binarySearch(arr, l, mid-1, x);
        else
            return binarySearch(arr, mid+1, r, x);
    }
    return -1;
}
```

Time Complexity		Space Complexity	
Binary	Linear	Binary	Linear
Iterative $\rightarrow O(\log n)$	$O(n)$	$O(1)$	$O(1)$
Recursive $\rightarrow O(1)$	$O(n)$	$O(1)$ or $O(\log n)$	$O(1)$
$O(\log n)$			

②4. $\rightarrow T(n)$

```
int bs(int arr, int l, int r, int x)
{
    while (l <= r) {
        int mid = (l+r)/2;
        if (arr[mid] == x) return mid;  $\rightarrow T(1)$ 
        return bs(arr, mid+1, r, x);
        else
            return bs(arr, l, mid-1, x);  $\rightarrow T(n/2)$ 
    }
    return -1;
}
```

$$T(n) = T(n/2) + T(n/2) + 1$$

$$\boxed{T(n) = T(n/2) + 1} \text{ Ans.}$$